

High Performance Computing for Information Retrieval

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 3980 - INSTRUMENT SOFTWARE AND SCIENCE DATA SYSTEMS





"Information retrieval (IR) is finding material (usually documents) of an unstructured nature (usually text) that satisfies an information need from within large collections (usually stored on computers)."

[Manning, Raghavan and Schutze (2008)]

- IR systems by the scale
 - web search
 - personal information retrieval
 - domain-specific search
- Challenging blend of science and engineering
- First major concept in IR: the inverted index

Inverted Index



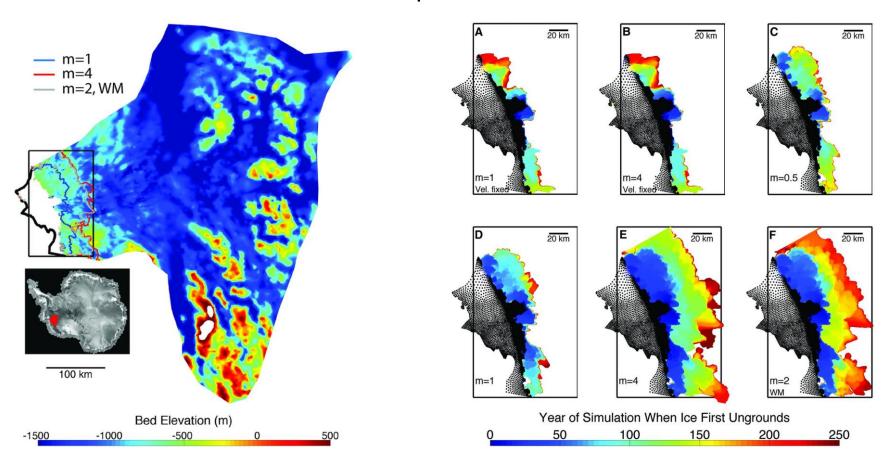
- Collect the documents to be indexed
- Tokenize the text
- Produce a list of normalized tokens

DICTIONARY **POSTINGS** Index the documents doc. freq. Xdocuments term 1:1, 4:6 forecasts 1:5, 2:1, 3:13, 4:10 home **Doc 1** | New | home | sales 2 2:17, 3:10-24 in increase 3:1 nverte 3 2:20, 3:27, 4:1 Doc 3 Increase in home 2 1:1, 2:6 es rise |new||home||s 2 2:12, 4:21 **Doc 4** | july sales 4 1:10, 2:6, 3:18, 4:15 1:16 top



Scientific Case Study

Automatic Classification and Interpretation of Polar Datasets

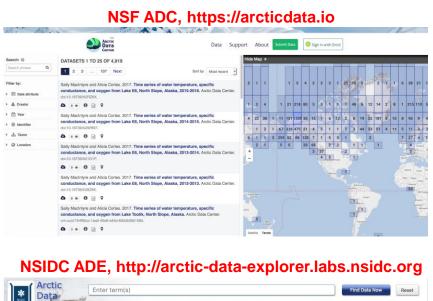


Thwaites Glacier bed topography derived from airborne radar data. [Joughin, Smith and Medley (2014)]



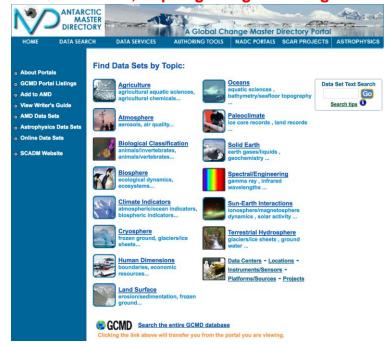
Scientific Case Study

Automatic Classification and Interpretation of Polar Datasets









[Burgess and Mattmann (2014)]



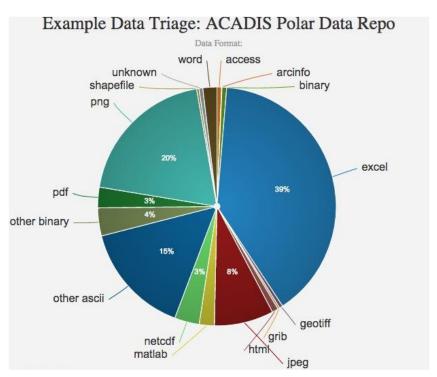
Scientific Case Study

Automatic Classification and Interpretation of Polar Datasets

- TREC Dynamic Domain: Polar Science
 - Data Triage
 - Web Crawl
 - Data Preparation
 - Dataset Characteristics
 - Querying the Data

http://trec-dd.org/2015/2015.html

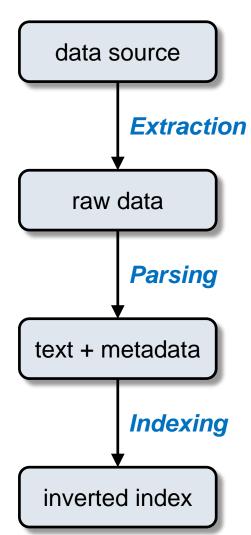
Dataset of 50,000+ crawled web pages, scientific data (HDF, NetCDF files, Grib files), zip files, PDFs, images and science code related to the polar sciences and available publicly from the NSF ADC, NASA AMD, and NSIDC ADE.



[Burgess, Mattmann, Totaro, McGibbney and Ramirez (2014)]

Indexing Data Flow





Extracting raw files from data sources (e.g., HDF, NetCDF, GRIB, PDF, etc)

Parsing operations:

- mimetype detection-
- language detection
- content handling
- metadata extraction

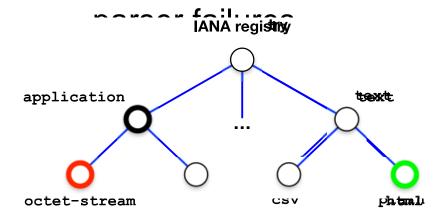
Indexing operations:

- tokenization
- stop words removal
- stemming

Main bottlenecks

- parsing of documents
- text analysis
- merging of segments

Detection complexity



Overview of the Research Problem

- Fast indexing and searching of Big Data
 - inverted index construction is a burdensome operation
- What is the H ap ecasystem?
 - MapReduce Dean and Gnemawar (2010)]
 - HDFS [Shafer, Rixner and Cox (2010)]
 - Ecosystem projects: Hive, Spark, Ambari, Pig, etc.
- Why not Hadoop?

 - Data Not Distributed

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- Hadoop performance [Jiang et al. (2010)] [Lin et al. (2012)] [Dong et al. (2014)]
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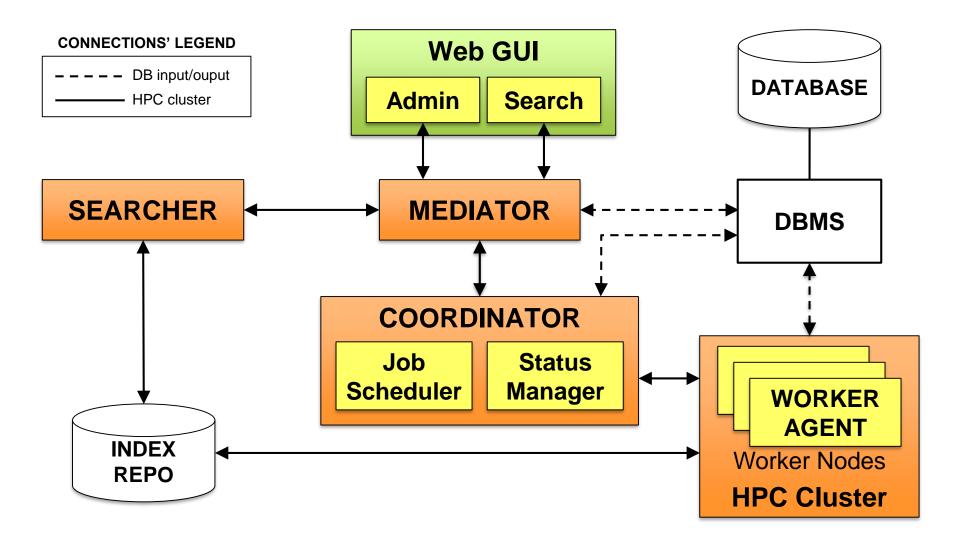


Research Statement and Hypotheses

- How to efficiently face the indexing problem?
 - Extract-Parse-Index in-memory pipelines
 - Parallel Technologies (e.g., GPU, Pthreads)
 - Enhancing de facto IR standard technologies
- ISODAC [Totaro, Bernaschi, Carbone and Cianfriglia (2016)]
 - 1. New high performance solution for indexing
 - 2. Only inverted indexes are written on disk
 - 3. Scalable according to hardware resources

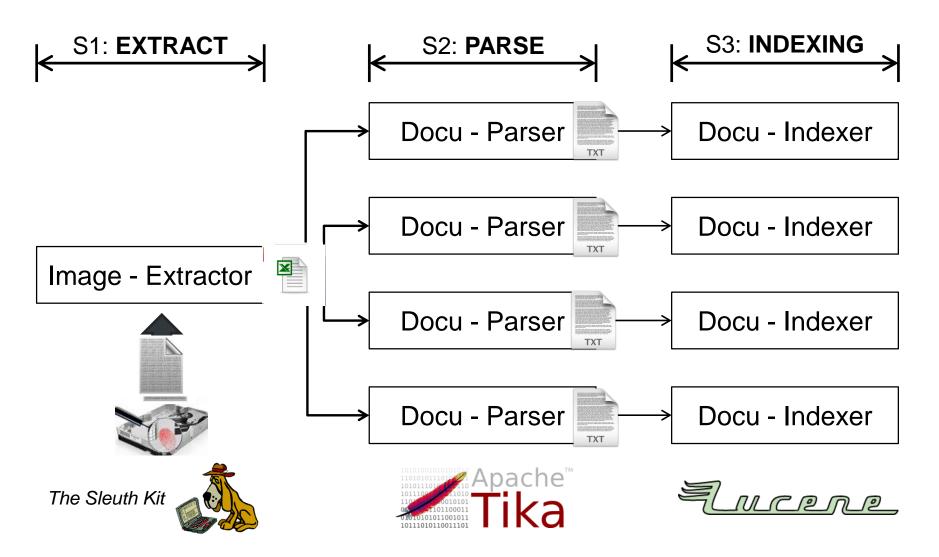


Software Architecture





EPI In-Memory Pipeline

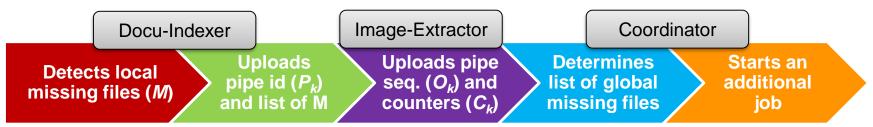




EPI In-Memory Pipeline

Recovery Procedure

 Provides resilience against both missing files (M) and parsing failures (rounds)



 Each global fileID can be calculated by using the following formula:

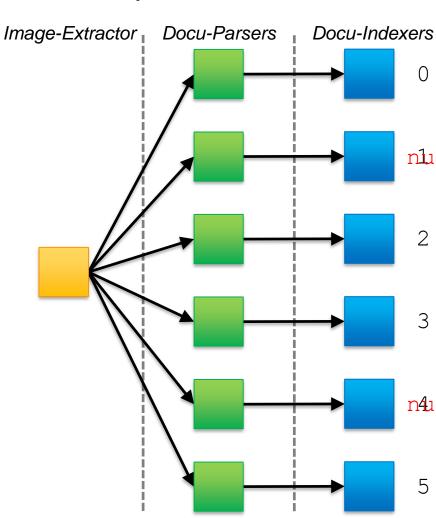
$$fileID = \begin{cases} M \times N[0] + O_k[0] & |rounds| = 0\\ sent[z] + (M - C_k[z]) \times N[z+1] + O_k[z+1] & |rounds| > 0 \end{cases}$$

$$(z = argmax_i | C_k[i] < M)$$

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EPI In-Memory Pipeline

Recovery Procedure



First round

$$1 = 0$$

$$nU11 N[0] = 6$$

$$O[0] = <0,1,2,3,4,5>$$

 $C[0] = \langle x_0, x_1, x_2, x_3, x_4, x_5 \rangle$

Second round

$$N[1] = 4$$
 $N[1] = 4$
 $C[1] = <0, \text{null}, 1, 2, \text{null}, 3>$
 $C[1] =$

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Optimizations: Text Analysis on GPU

- CUDA-based Text Analysis
 - Text analysis including tokenization, filtering and stop words removal processes
 - Exploits the computational power of GPU cards
 - Extends CLucene StandardAnalyzer with CUDA kernels that perform text analysis

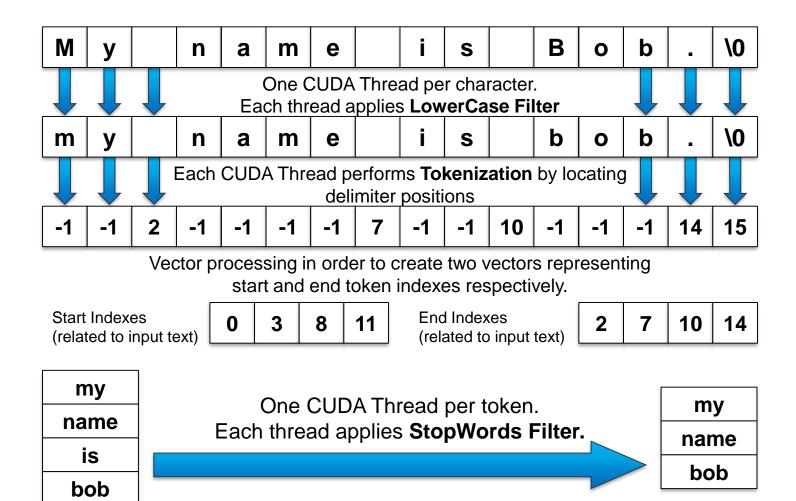




CLucene Libraries: http://clucene.sourceforge.net/

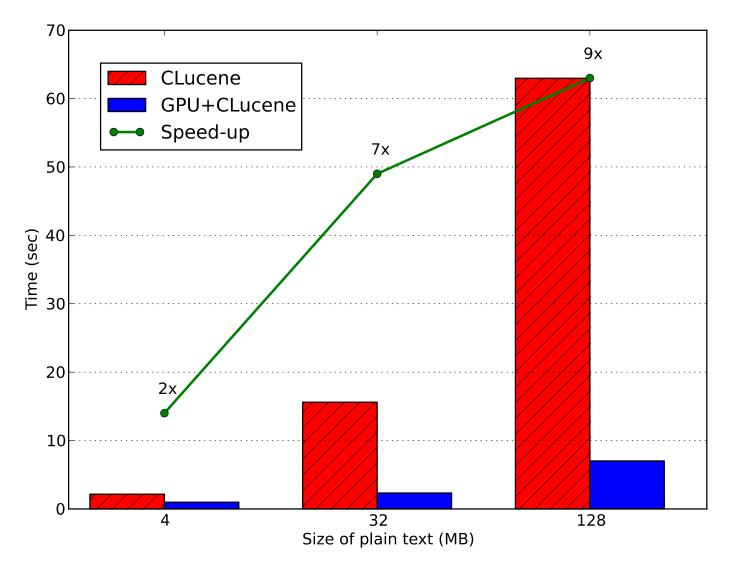
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Optimizations: Text Analysis on GPU





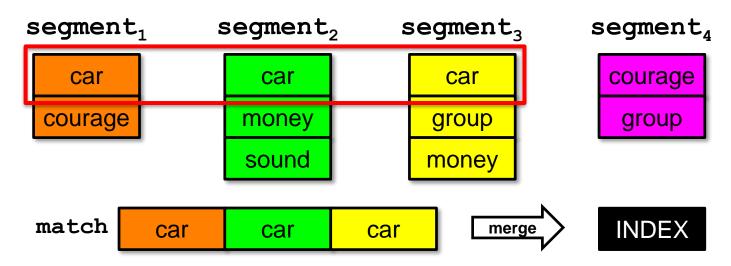
Optimizations: Text Analysis on GPU





Optimizations: Parallel Segment Merging

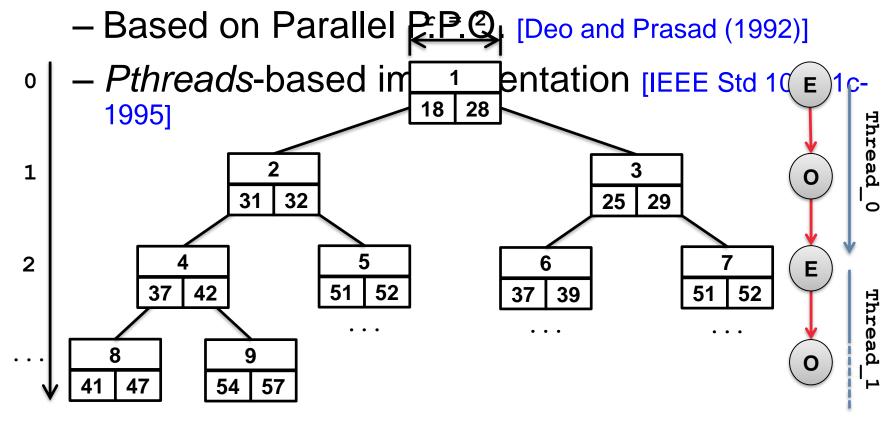
- A Lucene index may consist of multiple sub-indexes or segments
- Merging algorithms are usually based on priority queues





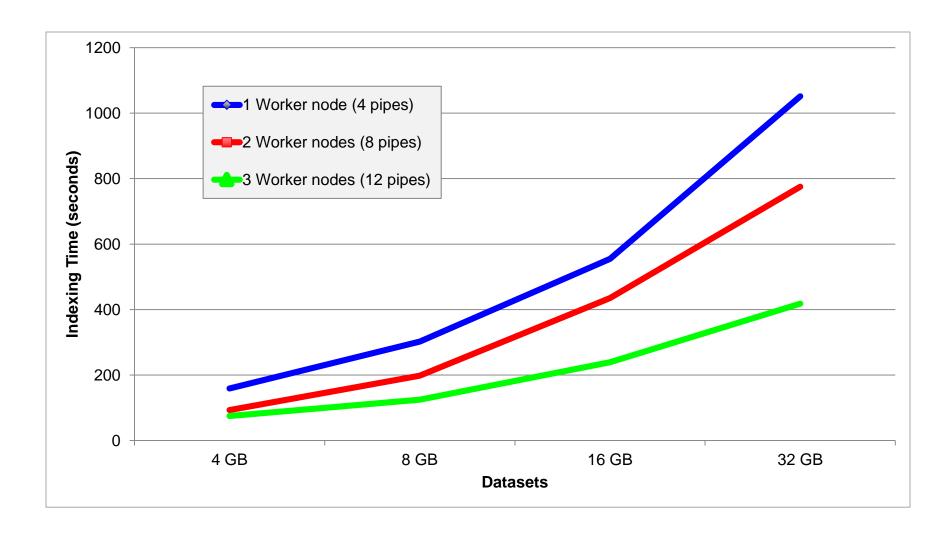
Optimizations: Parallel Segment Merging

 Perform in parallel multi-thread heapify and merging of extracted data



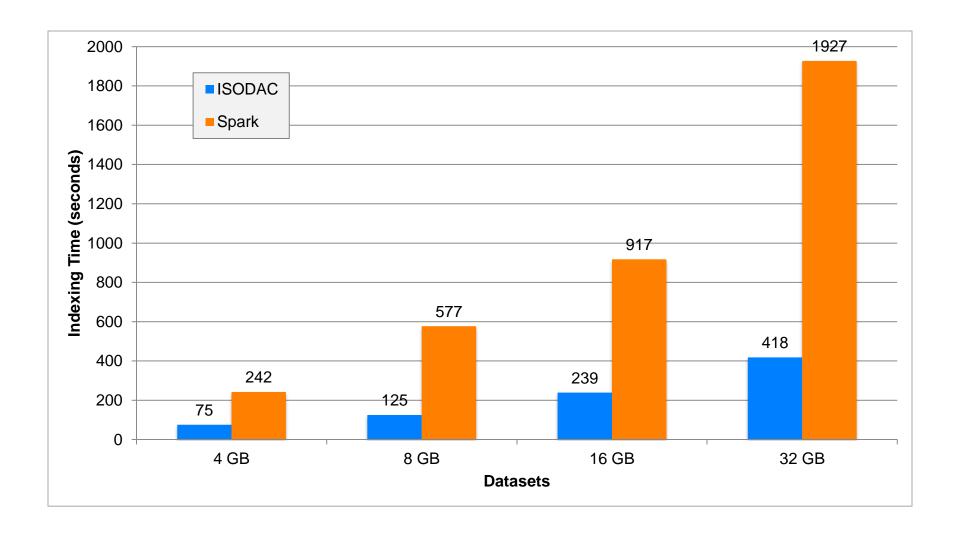


Evaluation: ISODAC performance





ISODAC vs. Spark Indexing





Fault Recovery

- Simulation of two kinds of failure:
 - Transient: node restarts after the failure
 - Permanent: node becomes permanently unavailable

Stress Pattern (16 GB dataset)	ISODAC		Spark	
	Time (secs)	Overhead	Time (secs)	Overhead
No Failure	239	-	917	-
Transient	387	61%	1131	45%
Permanent	415	73%	2542	177%



Metadata

- Illustrative Example
- Metadata Building Blocks
- Metadata A Model Perspective



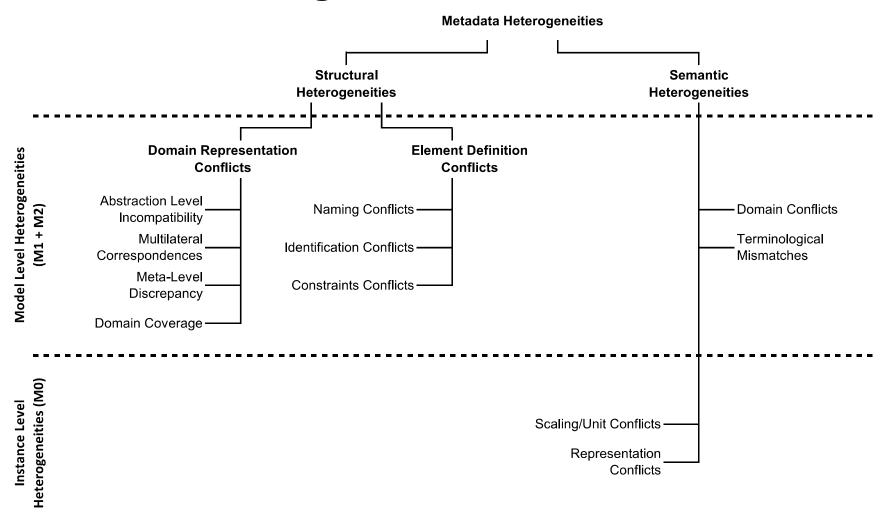
Metadata Interoperability

 Prerequisite for uniform access to media objects

"Metadata interoperability is a qualitative property of metadata information objects that enables systems and applications to work with or use these objects across system boundaries." [Haslhofer and Klas (2010)]

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Metadata Heterogeneities



Predominant heterogeneities have been originally identified by: [Sheth, Larson (1990)] [Ouksel, Sheth (1999)] [Wache (2003)] [Visser et al. (1997)]



Metadata Mapping

- Interoperability technique that subsumes:
 - schema mapping crosswalks
 - instance transformation functions

"Given a source schema $S^s \in S$ and a target schema $S^t \in \mathcal{S}$, each consisting of a set of schema elements, $e^s \in S^s$ and $e^t \in S^t$ respectively, a mapping $M \in \mathcal{M}$ is a directional relationship between a set of elements $e_i^s \in S^s$ and a set of elements $e_i^t \in S^t$." [Haslhofer and Klas (2010)]



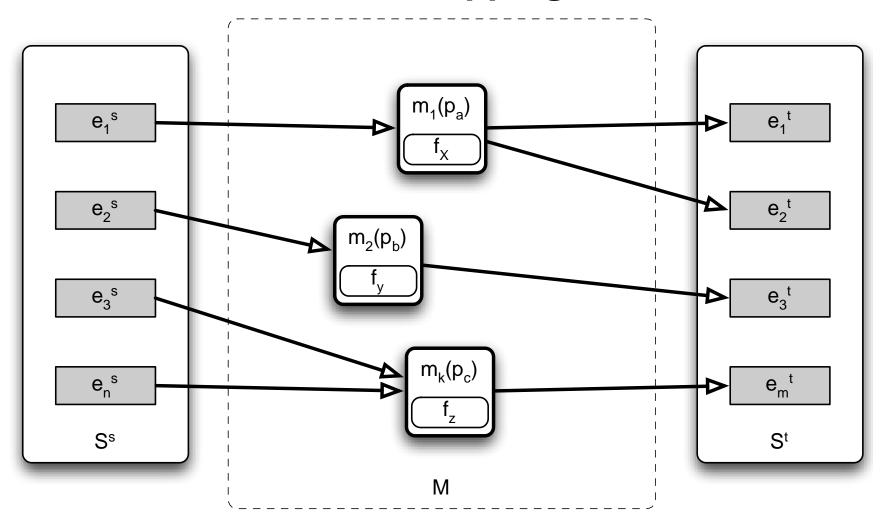
Mapping Relationship

$$m \in M \begin{cases} |m| & \text{Cardinality (e.g., 1:1, 1:n, n:1)} \\ p \in P & \text{Mapping expression} \\ f \in F & \text{Instance transformation function} \end{cases}$$

- Mapping expressions [Spaccapietra et al. (1992)]
 - Exclude: $I(e_i^s) \cap I(e_i^t) = \emptyset$
 - Equivalent: $I(e_i^s) \equiv I(e_j^t)$
 - Include: $\left(I(e_i^s) \subseteq I(e_j^t) \lor I(e_j^t) \subseteq I(e_i^s)\right)$
 - Overlap: $(I(e_i^s) \cap I(e_j^t) \neq \emptyset \land I(e_i^s) \nsubseteq I(e_j^t) \land I(e_j^t) \nsubseteq I(e_i^s))$



Elements of Metadata Mapping

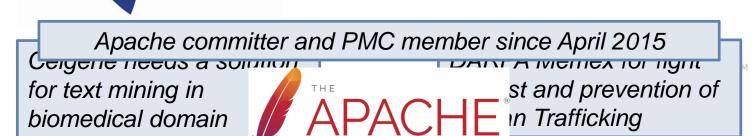




Other Group Projects at NASA JPL



Six months internship at NASA JPL working for **Computer Science for Data Intensive Applications (398M)** group (Mentor: Prof. Chris Mattmann)



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